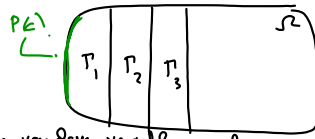


Expectations of discrete r.v.

Recall: $\{T_n\}_{n \in \mathbb{N}}$ is a quasi-partition of Ω if all T_n 's are events, $T_i \cap T_j = \emptyset$ for $i \neq j$, and $P(\Omega \cup T_n) = 1$



Recall: If X is a discrete random variable and $\{T_n\}$ is a quasi-partition such that X is constant on each T_n with $X = x_n$. Then X is integrable iff $\sum_{n=1}^{\infty} |x_n| P(T_n) < \infty$ and then $E[X] = \sum_{n=1}^{\infty} x_n P(T_n)$

Theorem: Let $X, Y: \Omega \rightarrow \mathbb{R}$ be two discrete random variables.

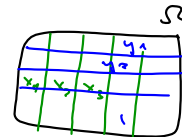
- Then
- (i) X is integrable iff $|X|$ is integrable.
 - (ii) If X, Y are integrable and $a, b \in \mathbb{R}$, then $aX + bY$ is integrable and $E[aX + bY] = aE[X] + bE[Y]$
 - (iii) If $|X| \leq |Y|$ and Y is integrable, then X is integrable.
 - (iv) If X, Y are integrable and $X \leq Y$, then $E[X] \leq E[Y]$
 - (v) If X is integrable, then $E[X] \leq E[|X|]$

Proof: (i) We have X integrable $\Leftrightarrow \sum_{i=1}^{\infty} |x_i| P(T_i) < \infty$
 $|X|$ integrable $\Leftrightarrow \sum_{i=1}^{\infty} |x_i| P(T_i) < \infty$

(ii) Let $\{x_i\}$ be the points that X takes with prob > 0
 --- $\{y_j\}$ --- " --- Y --- " ---

Define a quasi-partition $T_{i,j}$:

$$T_{i,j} = \{\omega : X(\omega) = x_i \text{ and } Y(\omega) = y_j\}$$



Assume X and Y are integrable. Then $aX + bY$ is integrable because:

$$\sum_{i,j} |ax_i + by_j| P(T_{i,j}) \leq \sum_{i,j} (|a||x_i| + |b||y_j|) P(T_{i,j})$$

$$= |a| \sum_{i,j} |x_i| P(T_{i,j}) + |b| \sum_{i,j} |y_j| P(T_{i,j}) < \infty$$

Also

$$E[aX + bY] = \sum_{i,j} (ax_i + by_j) P(T_{i,j})$$

$$= a \sum_{i,j} x_i P(T_{i,j}) + b \sum_{i,j} y_j P(T_{i,j}) = aE[X] + bE[Y]$$

(iii) Assume $|X| \leq |Y|$ and Y is integrable:

$$\sum_{i,j} |x_i| P(T_{i,j}) \leq \sum_{i,j} |y_j| P(T_{i,j}) < \infty$$

Hence X is integrable.

(iv) Assume X, Y integrable and $X \leq Y$. Must show $E[X] \leq E[Y]$

$$E[X] = \sum_{i,j} x_i P(T_{i,j}) \leq \sum_{i,j} y_j P(T_{i,j}) = E[Y]$$

(v) Assume X integrable. Since $X \leq |X|$, (iv) gives us

$$E[X] \leq E[|X|]$$

Expectation for general random variables

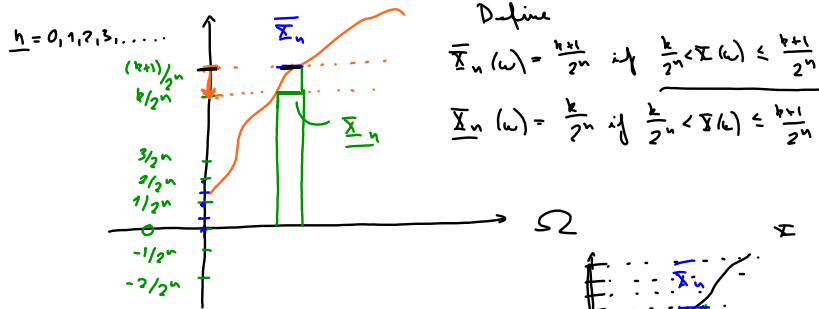
X r.v. Problem: How to define $E[X]$?

Idea: Find sequences $\{\bar{X}_n\}$ and $\{\underline{X}_n\}$ of discrete random variables such

$$\underline{X}_n \leq X \leq \bar{X}_n$$

$$\underline{X}_n \nearrow X, \bar{X}_n \searrow X \quad \text{and} \quad \lim_{n \rightarrow \infty} E[\bar{X}_n] = \lim_{n \rightarrow \infty} E[\underline{X}_n]$$

Define $E[X] = \lim_{n \rightarrow \infty} E[\bar{X}_n] = \lim_{n \rightarrow \infty} E[\underline{X}_n]$



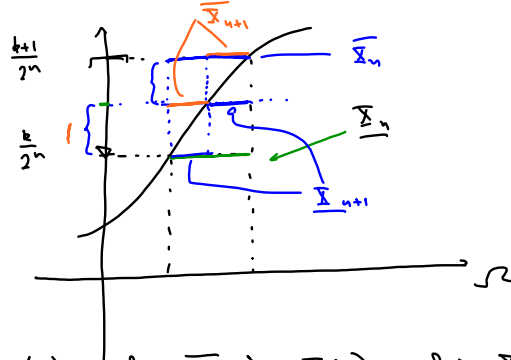
Observations:

(i) \bar{X}_n and \underline{X}_n are random variables and

$$\underline{X}_n(\omega) < X(\omega) \leq \bar{X}_n(\omega)$$

$$\text{and } \bar{X}_n(\omega) - \underline{X}_n(\omega) = \frac{1}{2^n}$$

(ii) The sequence $\{\bar{X}_n\}$ is decreasing and $\{\underline{X}_n\}$ is increasing (i.e. $\bar{X}_{n+1}(\omega) \leq \bar{X}_n(\omega)$ and $\underline{X}_{n+1}(\omega) \geq \underline{X}_n(\omega)$)

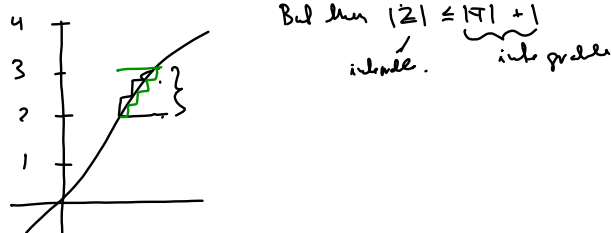


$$(iii) \quad \lim_{n \rightarrow \infty} \bar{X}_n(\omega) = X(\omega) = \lim_{n \rightarrow \infty} \underline{X}_n(\omega)$$

(iv) The approximations \bar{X}_n and \underline{X}_n are either all integrable or all nonintegrable.

Proof: It suffices to prove that if 1 and 2 are two approximations and 1 is integrable, then 2 is also integrable.

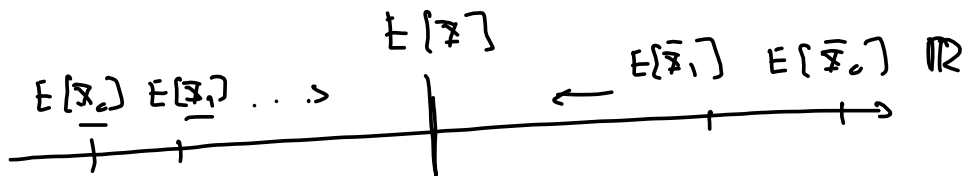
$$\text{Since 1 and 2 are approximations, } |1-2| \leq 1.$$



We have $E[\bar{X}_n]$ is decreasing

— — — $E[\underline{X}_n]$ is increasing

$$E[\bar{X}_n] - E[\underline{X}_n] = E[\bar{X}_n - \underline{X}_n] \leq E[|\bar{X}_n - \underline{X}_n|] = \frac{1}{2^n} \cdot C$$



We have shown that $\lim_{n \rightarrow \infty} E[\underline{X}_n] = \lim_{n \rightarrow \infty} E[\bar{X}_n]$

Definition: Let X be a random variable. We say that X is integrable if \bar{X}_0 is integrable (which is the same as saying $\bar{X}_n, \underline{X}_n$ are integrable for all n).

In this case, we define the expectation as

$$E[X] = \lim_{n \rightarrow \infty} E[\bar{X}_n] = \lim_{n \rightarrow \infty} E[\underline{X}_n].$$

Theorem: Let $X, Y: \Omega \rightarrow \mathbb{R}$ be random variables. Then

- (i) If $X = Y$ a.s. (i.e. $P(\omega: X(\omega) \neq Y(\omega)) = 0$) and Y is integrable, then X is integrable and $E[X] = E[Y]$.
- (ii) If $|X| \leq |Y|$ and Y is integrable, then X is integrable.
- (iii) If X is integrable and $a \in \mathbb{R}$, then aX is integrable and $E[aX] = aE[X]$.
- (iv) If X, Y are integrable, then $X+Y$ is integrable and $E[X+Y] = E[X] + E[Y]$.
- (v) If $X \geq 0$ a.s., then $E[X] \geq 0$.
- (vi) If X, Y are integrable and $X \leq Y$, then $E[X] \leq E[Y]$.

Proof: (i) $X = Y$ a.e., Y integrable.

$$\underline{X \text{ is integrable}} \Leftrightarrow \bar{X}_n \text{ is integrable} \Leftrightarrow \sum \left| \frac{k+1}{2^n} \right| P \left[\frac{k}{2^n} < \overset{Y}{X}(\omega) \leq \frac{k+1}{2^n} \right] < \infty$$

$$\Leftrightarrow \sum \left| \frac{k+1}{2^n} \right| P \left[\frac{k}{2^n} < Y(\omega) \leq \frac{k+1}{2^n} \right] < \infty \Rightarrow Y \text{ is integrable.}$$

$$E[\bar{X}_n] = \sum \frac{k+1}{2^n} P \left[\frac{k}{2^n} < X(\omega) \leq \frac{k+1}{2^n} \right]$$

$$= \sum \frac{k+1}{2^n} P \left[\frac{k}{2^n} < Y(\omega) \leq \frac{k+1}{2^n} \right] = E[\bar{Y}_n]$$

$$\text{Hence } E[X] = \lim_{n \rightarrow \infty} E[\bar{X}_n] = \lim_{n \rightarrow \infty} E[\bar{Y}_n] = E[Y]$$

(ii) Assume $|X| \leq |Y|$ and Y is integrable. Show that X is integrable. We have

$$|\bar{X}_0| - 1 \leq |X| \leq |Y| \leq |\bar{Y}_0| + 1$$

Hence $|\bar{X}_0| \leq |\bar{Y}_0| + 2$, and hence \bar{X}_0 is integrable. By definition X is integrable.

(iii) X is integrable. Want to show that aX is integrable and that $E[aX] = aE[X]$.

Note that. $|aX| = |a||X| \leq |a|(|\bar{X}_0| + 1)$

Hence aX is integrable by (ii). integrable

Let $Z = aX$. Then

$$|\bar{Z}_n - a\bar{X}_n| = |(\bar{Z}_n - Z) + (aX - a\bar{X}_n)|$$

$$\leq |\bar{Z}_n - Z| + |a||X - \bar{X}_n| \leq \frac{1}{2^n} + |a| \frac{1}{2^n} = \frac{|a|+1}{2^n}$$

$$\text{Hence } \begin{matrix} E[Z] & aE[X] \\ \uparrow & \uparrow \\ |E[\bar{Z}_n] - aE[\bar{X}_n]| = |E[\bar{Z}_n - a\bar{X}_n]| \end{matrix}$$

$$\leq E[|\bar{Z}_n - a\bar{X}_n|] \leq \frac{|a|+1}{2^n} \rightarrow 0$$

The two limits have to be the same, i.e.

$$E[Z] = aE[X].$$